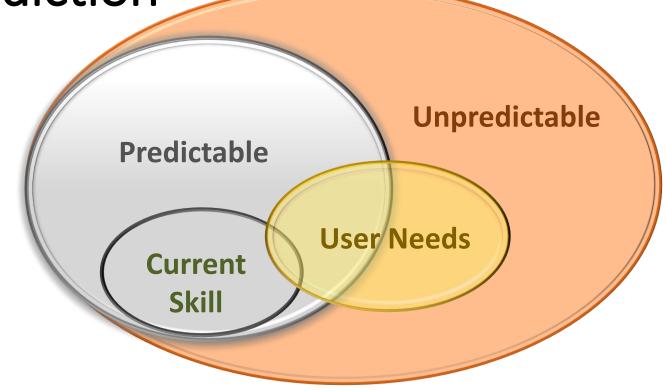
Subseasonal and Seasonal Precipitation:

from Predictability to Prediction

Kathy Pegion
George Mason University

Thanks to: Emily Becker,

NMME Team, SubX Team





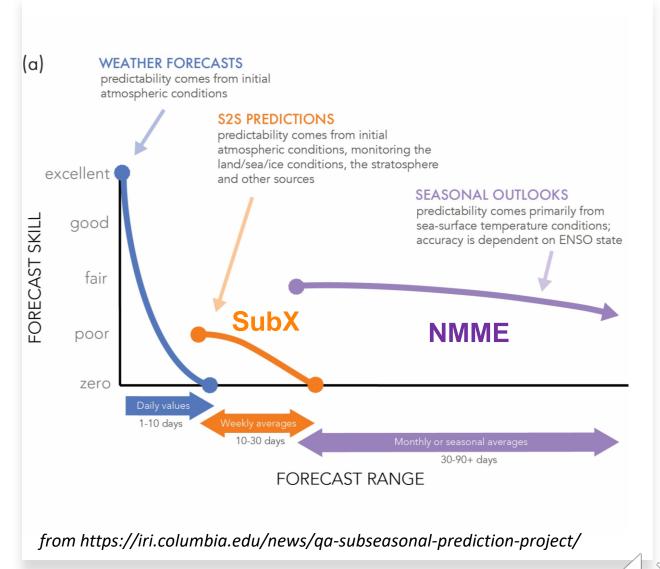
How good are our predictions now?

What is the limit of predictability?

How can we make better predictions?



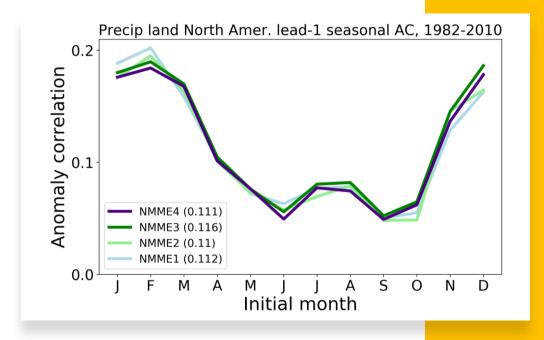
# How good are our predictions now?

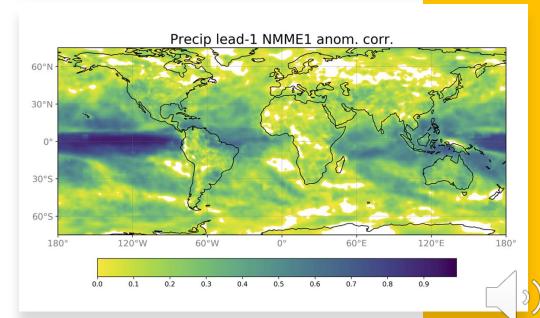


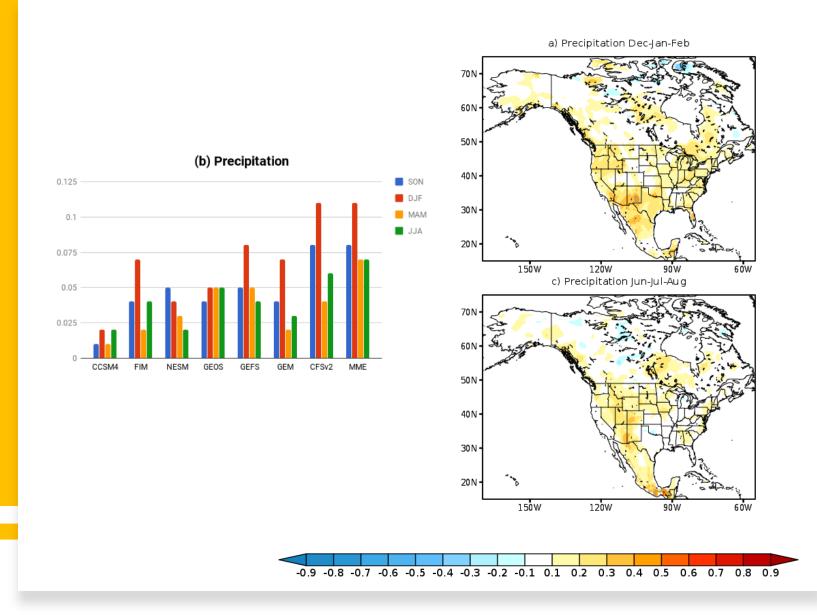
2)

### North American Multi-model Ensemble (NMME)

- Varies in Space and Time
- Low over North America
- No significant improvement



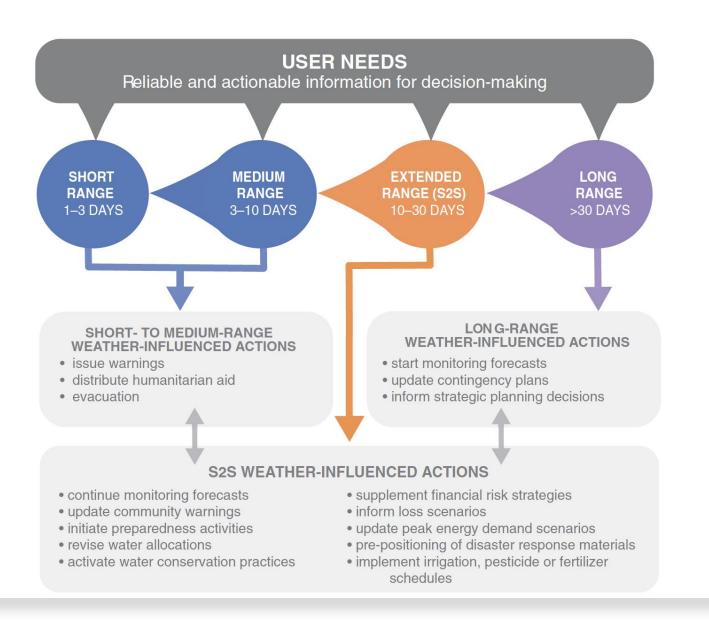




### The Subseasonal Experiment (SubX)

- Varies in Space and Time
- Low over North America

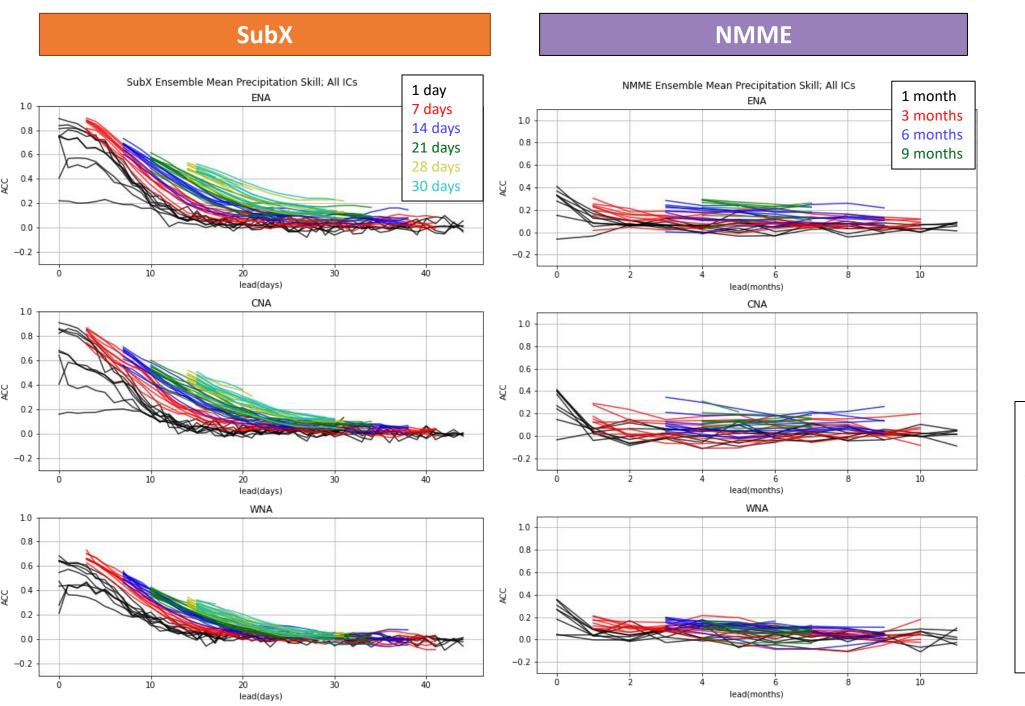




This timescale is used for planning

A flexible approach to assessing skill in space and time





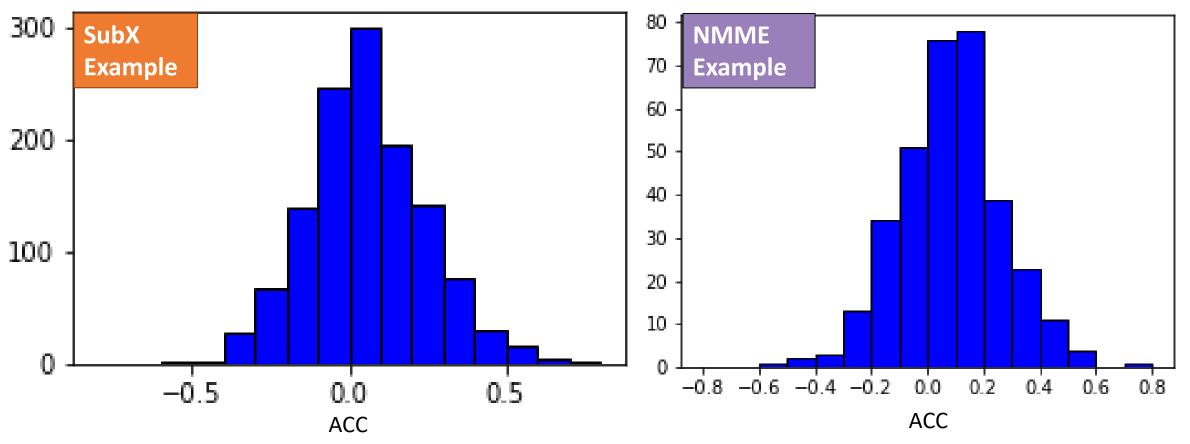


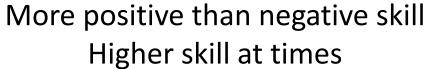
Better skill when averaged in space and time

On average skill is still relatively low

Explore other optimal approaches

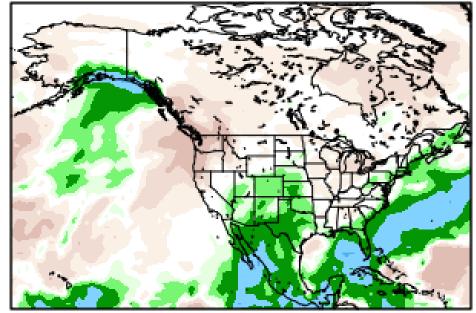
#### Forecasts of Opportunity







#### SubX Week 3-4 Total Precipitation Anomalies (mm): Valid 2 weeks ending OCT 19 MME (63 Ensemble Members)



from Pegion et al. 2019, BAMS



There is anecdotal evidence of useful forecasts that would not be skillful using traditional skill measures.

Can we find ways to identify, use, and quantify them?



# What is the limit of predictability?

Function of signal to noise:

$$P = F\left(\frac{\sigma_S^2}{\sigma_N^2}\right)$$

Perfect model predictability

Assume: Model is perfect, only source of error is initial condition uncertainty

Signal = *estimated* by the ensemble mean

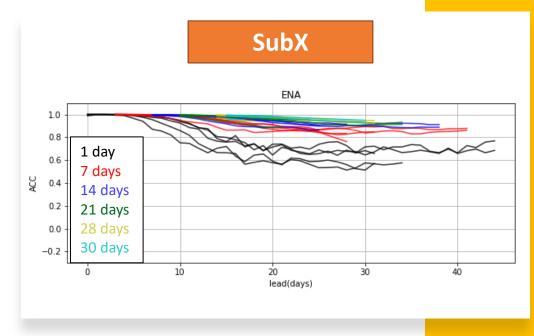
Noise = *estimated* by the ensemble spread

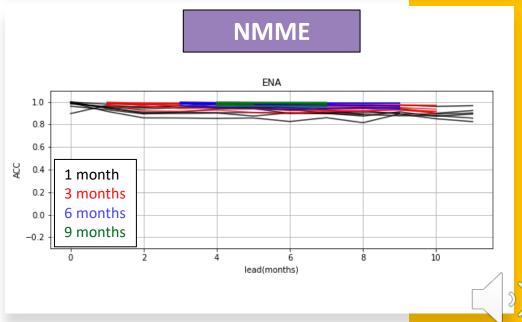
## We do not know the upper limit of skill

Unrealistic estimates

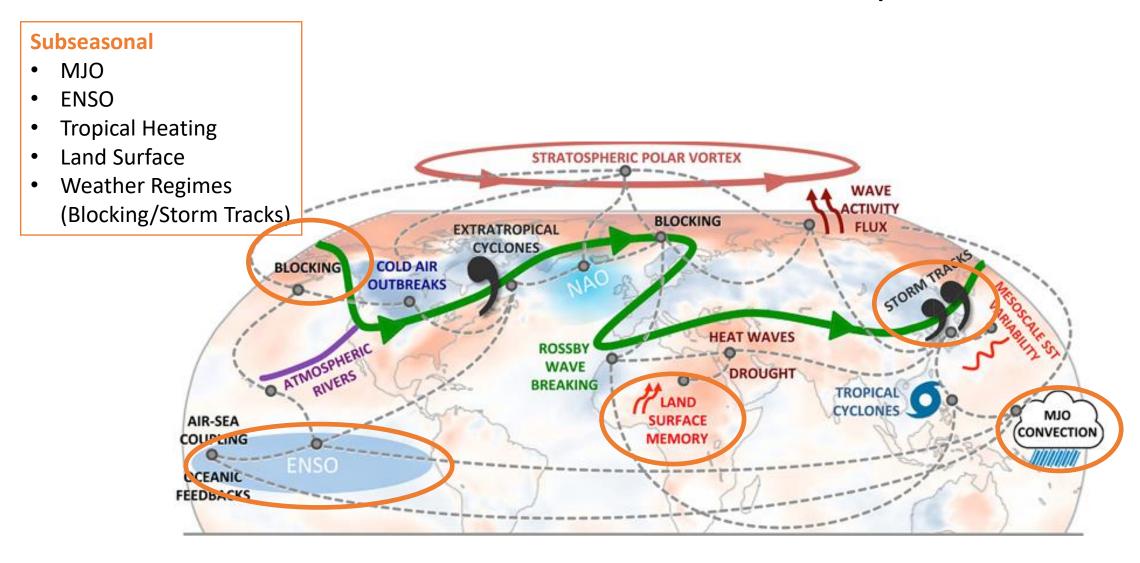
Noise is large at these timescales

Understand predictability by understanding signal





#### Understand Sources & Impacts





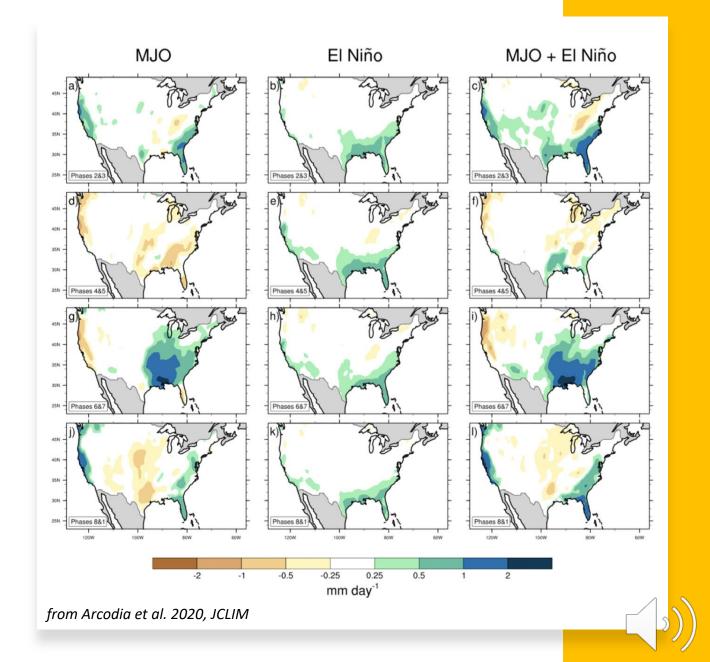
#### MJO and ENSO

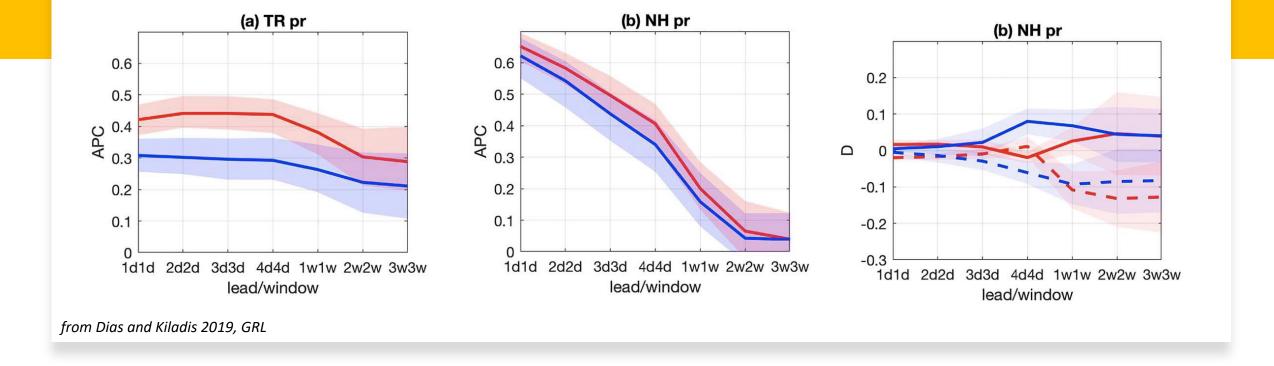
MJO and ENSO interact to impact CONUS precipitation

Still have fundamental MJO errors in our models (propagation, initiation)

How well do our models represent these interactions and impacts on precipitation?

What about other MJO interactions (e.g. QBO, ARs)?





#### **Tropical Heating**

Better tropical precip skill at short lead-times leads to better subseasonal precip skill in NH.

Skill is still relatively low

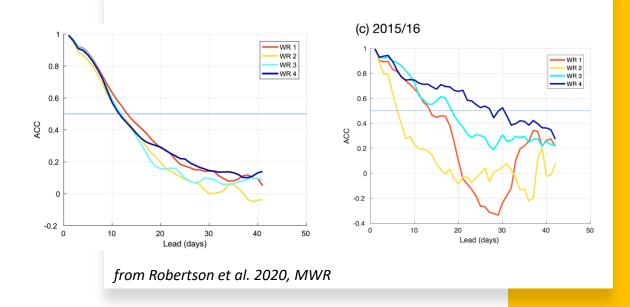
How well could we predict tropical precipitation? How much CONUS skill could we get from the tropics

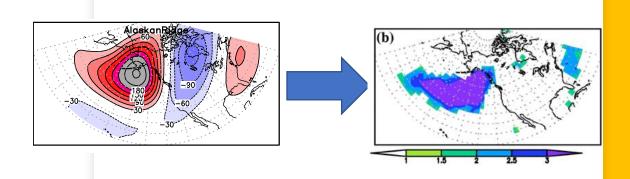
#### Weather Regimes

Pacific North America weather regimes predicted to ~15 days and longer in certain cases.

There is a relationship between these regimes and CONUS precipitation.

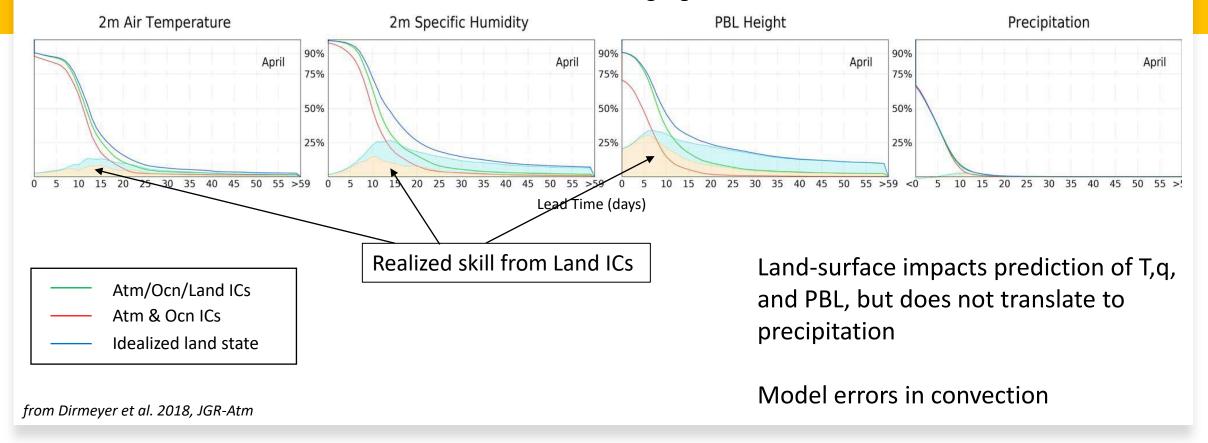
Can we represent that relationship and realize this predictability?





from Amini and Straus 2019, Clim Dyn

#### Fraction of Land Area Showing Significant Skill



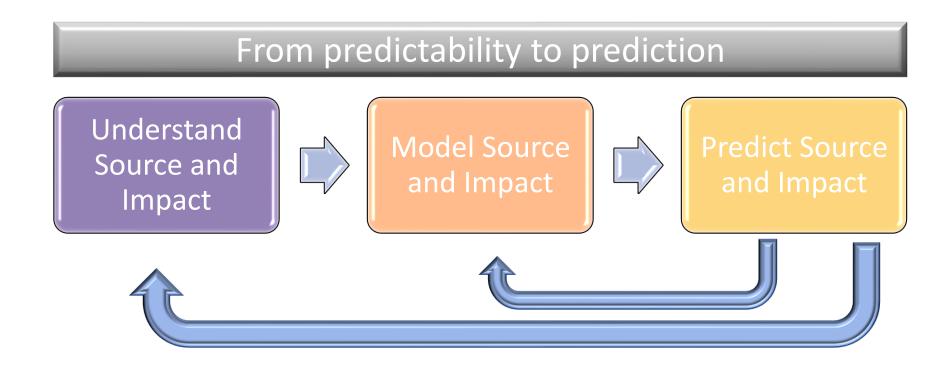
#### **Land Surface**

Can we fix this in our prediction models?

If we fix it, how much skill improvement can we get?



#### How can we make better predictions?





#### Summary

#### How good are our predictions now?

- Currently skill is low on average
- We can't measure all potentially useful forecasts
- Traditional skill metrics may be misleading

#### What is the limit of predictability?

- We don't know for sure
- There are forecasts of opportunity

#### How can we make better predictions?

- Focus on forecasts of opportunity
- Better understand and model impacts of the sources of predictability
- Acknowledge and quantify uncertainty

